

Navigating non-randomized comparisons: from pitfalls to practice

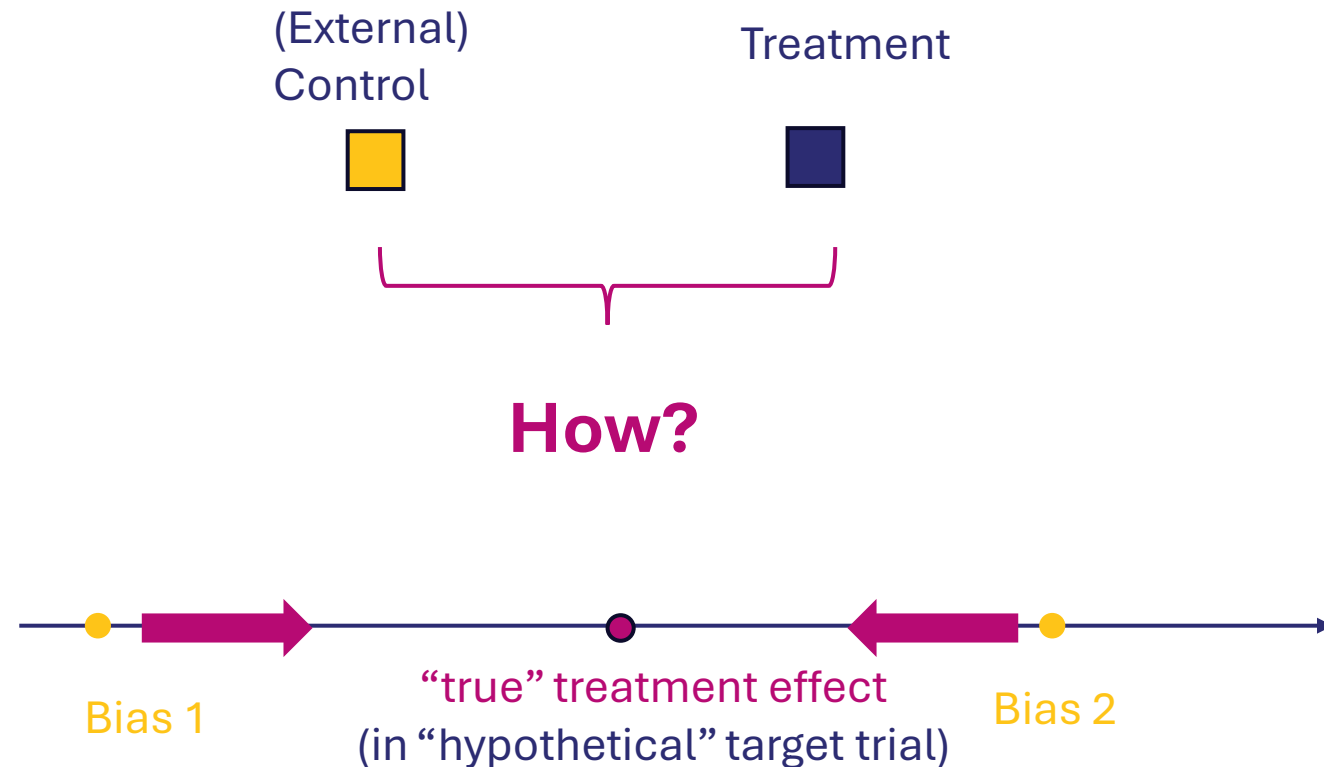
Workshop on the use of external data; EMA
Amsterdam, 3rd November 2025

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Acknowledging contributions of several colleagues at EFSPi

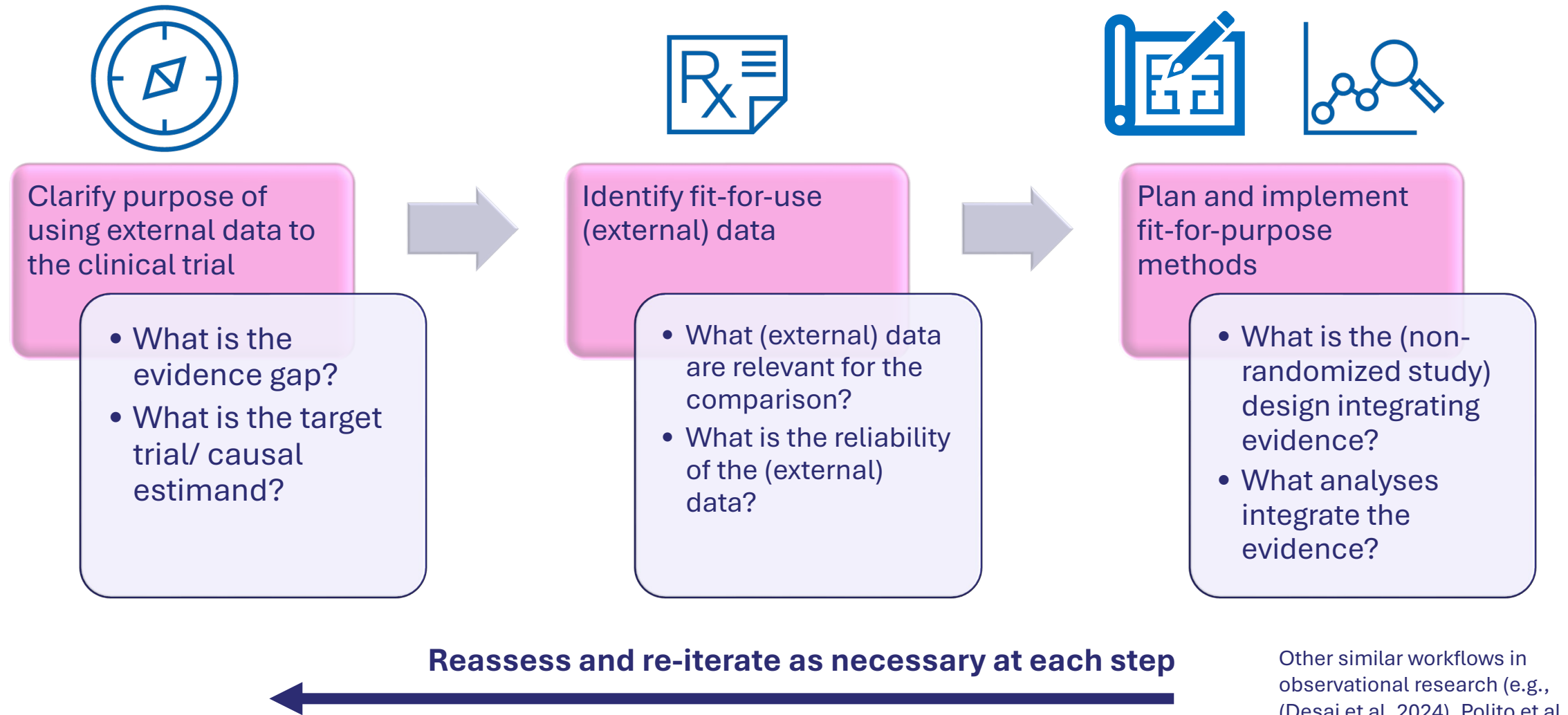
Navigating the pitfalls: reducing or mitigating biases

- **Added value:** Fair comparison of a treatment investigated in one arm in a trial to a control treatment investigated in another (external) source.
- **Pitfalls:** systematic biases* (e.g., confounding, selection, outcome assessment)
- **Methods** (design/analyses) in observational research or integrated evidence **can reduce bias**. Feasibility of these methods and trust in their results require,
 - Sufficient information on subjects and sources
 - Explicit assumptions
 - Several diagnostic checks and sensitivity analyses



*Potential sources of bias (Sterne et al 2019)

The journey of leveraging external data



Other similar workflows in observational research (e.g., (Desai et al, 2024), Polito et al (2024), Dahabreh et al (2024))

Case study in a rare blood disorder

Pivotal



- What is the effect of switching to iptacopan versus remaining on anti-C5 on hematological response in adult PNH patients **already on anti-C5**



- **APPLY-PNH (pivotal Ph3)** Double-blinded randomized, active control trial (iptacopan vs. anti-C5)

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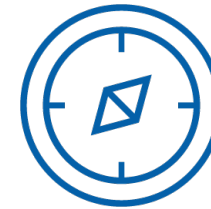


- What is the effect of iptacopan on hematological response in adult PNH patients **naive to anti-C5 treatment?**



- **APPOINT-PNH (pivotal Ph3)** single arm trial

Supportive



What is the effect of iptacopan versus anti-C5 on hematological response in adult PNH patients **naive to anti-C5 treatment?**

Case study, data and design



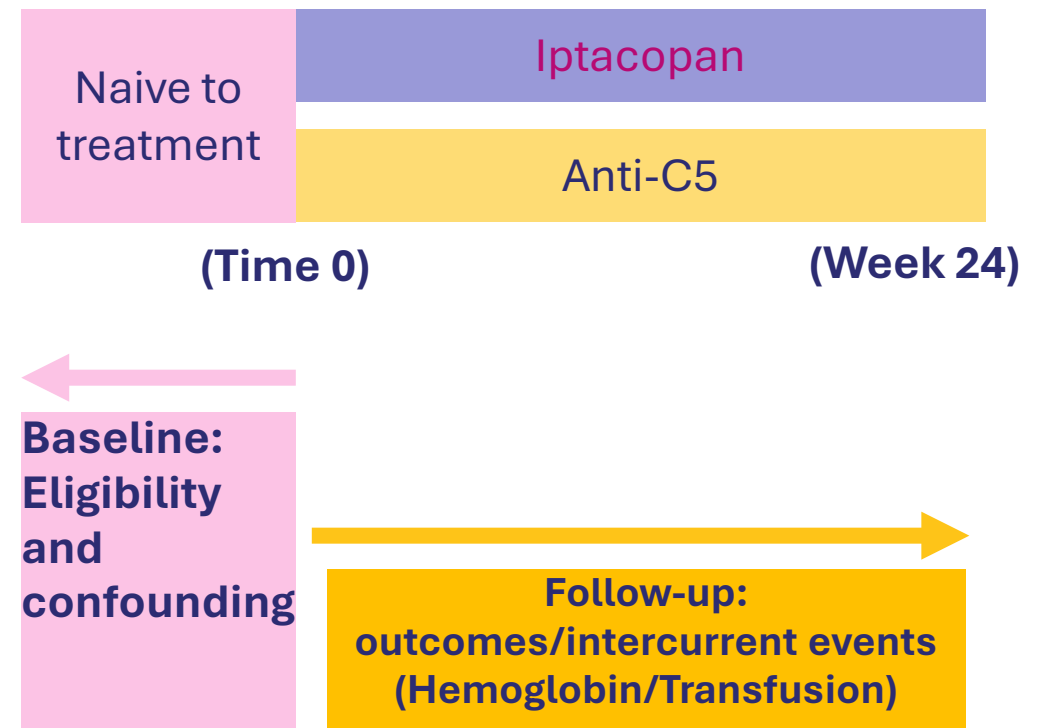
What is the effect of iptacopan versus anti-C5 on hematological response in adult PNH patients **naive to anti-C5 treatment**?



- **Single arm** - APPOINT-PNH (iptacopan)
- **External control** – APPEX, a PNH longitudinal “registry” from extracted medical chart review (anti-C5)



Target trial emulation



PNH: Paroxysmal Nocturnal Hemoglobunuria

Source: APPOINT-PNH, Peffault de Latour R. et al (2024); APPEX, Holt et al (2025)

Case study, summary of methods



Sources of bias	Mitigation
Selection of patients	- Emulate the target trial to align populations at time-0 prioritizing essential eligibility criteria
Confounding	- Identify prognostic variables by expert opinion, collect them in data sources¹ - Fit propensity score model(s) using prognostic variables,² balance variables using weighting³ - Estimate the treatment effect (ATT) using a doubly robust estimator⁴
Misaligned outcome assessment schedule	Sensitivity analyses to different definitions of visit-windows

Source: APPEX, Holt et al (2025)

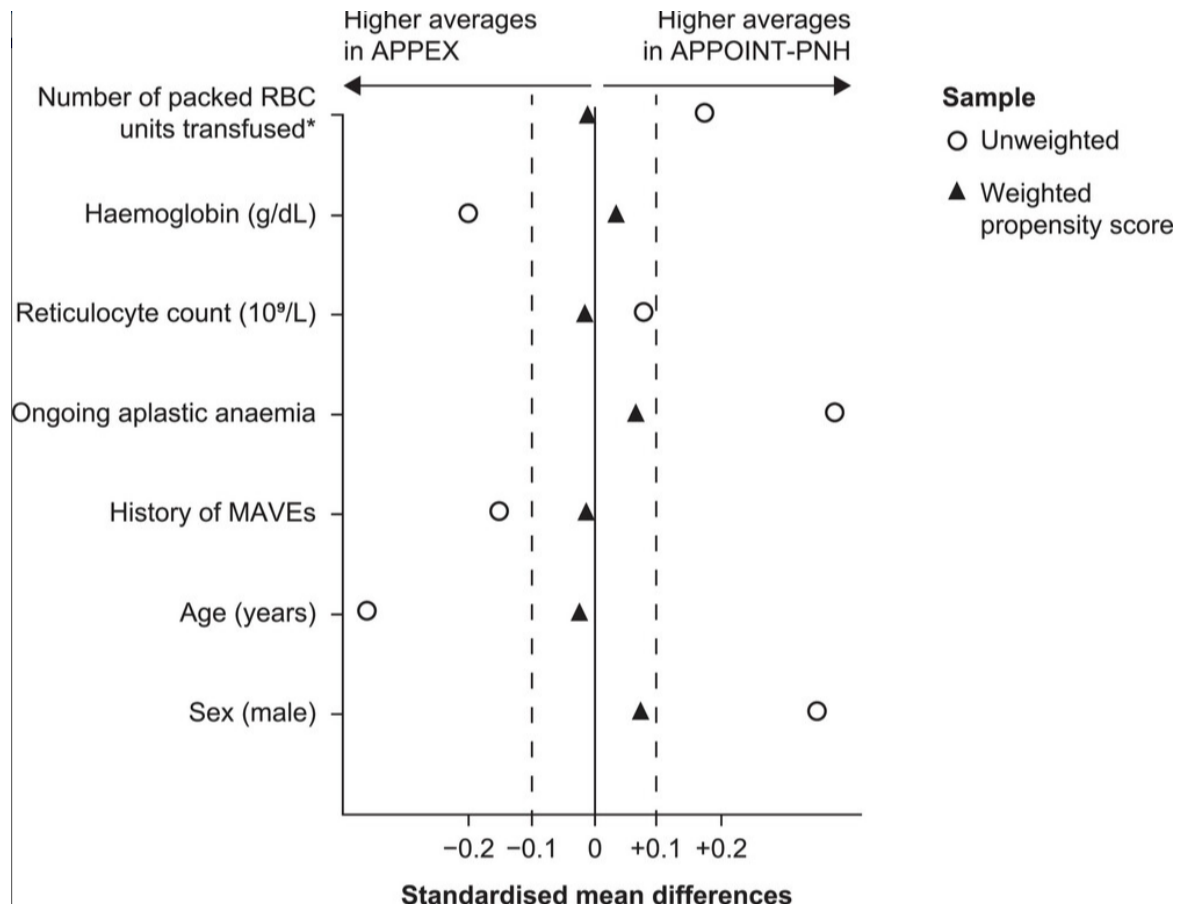
1 - evaluate the plausibility of exchangeability assumption of no unmeasured (residual) confounding

2 - evaluate the plausibility of the positivity assumption (e.g., distribution overlap of propensity scores)

3- evaluate the impact of different models on balance before/after weighting using standardized mean differences and statistical tests (Austin and Stuart 2015)

4 - ATT: average treatment on the treated weight by odds, AIPW: augmented inverse probability treatment weights (Mercatanti 2014, Chernozhukov et al 2018)

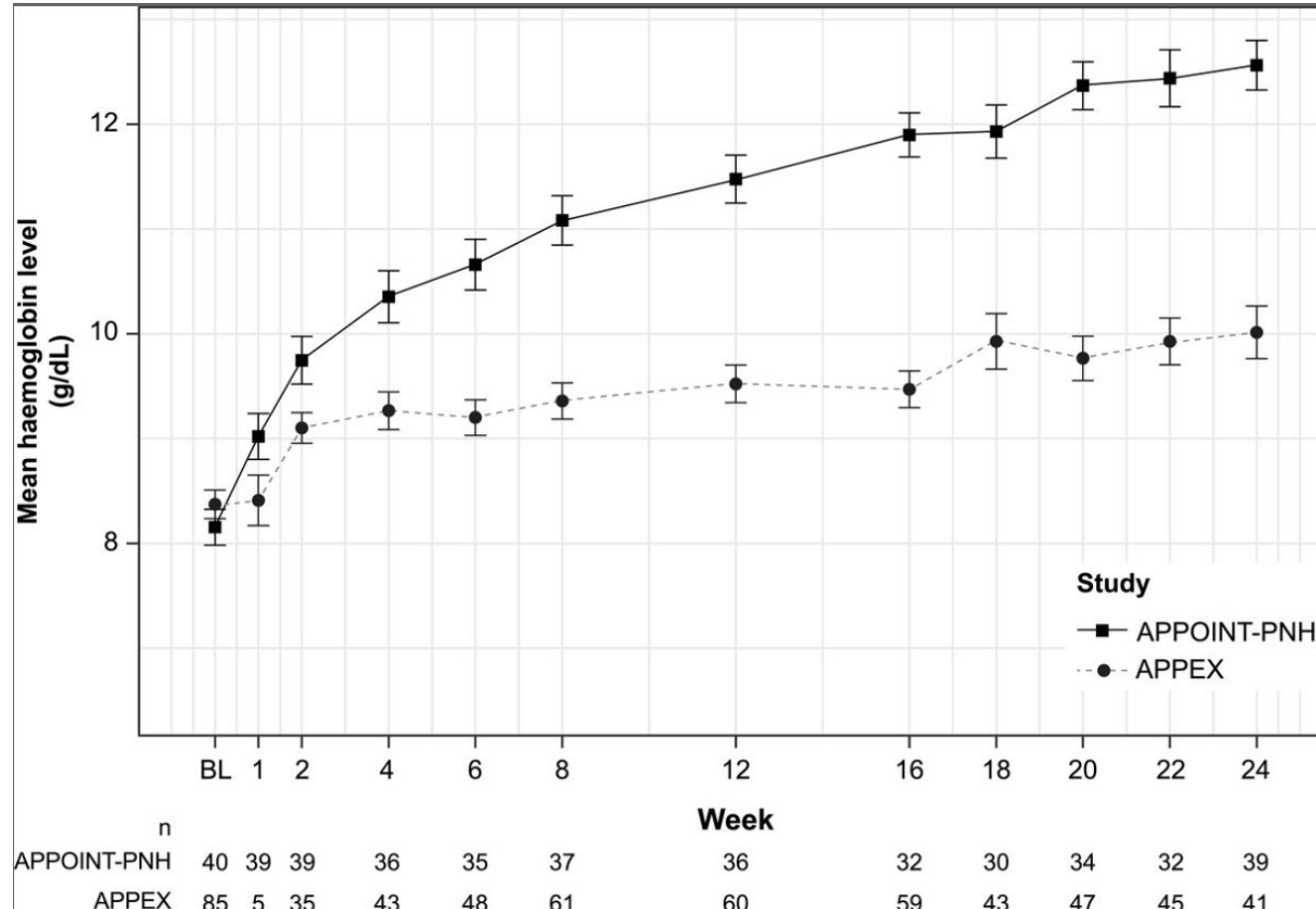
Diagnostics – potential confounding bias



- **Potential confounding bias:** imbalance between APPEX and APPOINT-PNH on all but one prognostic factor (Reticulocyte count). The imbalance did not systematically favor one group
- **Mitigation:** methods reduced potential confounding. After adjustment, all prognostic factors are balanced

Source: Holt et al (2025)

Select results (exploratory)



Source: Holt et al (2025)

- **Added value:** fair comparison of treatments and granularity of the treatment effect over time
- APPOINT-PNH (single arm) shows immediate improvement in mean hemoglobin after treatment with iptacopan, sustained until week 24
- Adjusted APPEX (external control) show small improvement after anti-C5 treatment

In summary, when making a comparison not protected by randomization

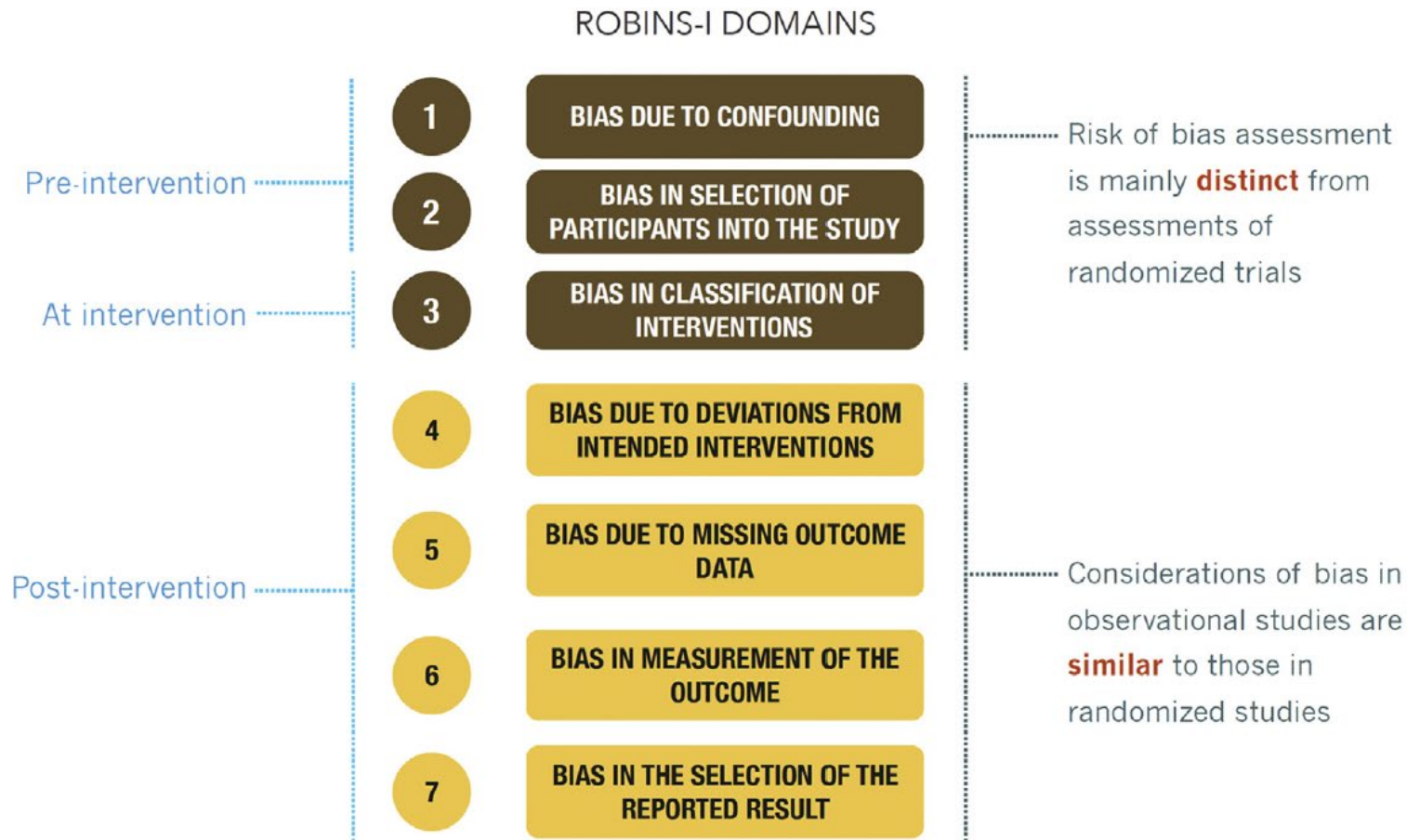
- Design and analytical strategies can **reduce bias** using eligibility and prognostic covariates with **explicit assumptions** on (dis)similarity of data sources
- **Quantitative bias/sensitivity analyses** can assess the impact of assumption violation or residual source of bias on the treatment effect estimate AND **diagnostics** can check plausibility of some assumptions
- **Challenge:** when and what to pre-specify in methods? What have we learned from similar considerations in guidance for meta-analyses (EMA 2001) and model informed drug development (ICH M15))?

Select references

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- ICH M15 (2024) Guideline on general principles for model-informed drug development (draft)

Back-up

Navigating non-randomized comparisons, naming the pitfalls or sources of bias



- confounding, or failure to account for a common cause of treatment and outcome (e.g., frailty (external control) at baseline > single arm)
- selection, erroneously conditioning on a common effect of treatment and outcome (e.g., immortal time in single arm)

Case study, minding the bias with TTE

Attributes	Target trial	Emulation in APPOINT (N=40)	Emulation in APPEX (N=85)
Population	Patients with PNH, naive to anti-C5, and with hemoglobin level < 10g/dL	More selection criteria than target trial [Sites primarily in Asia 2021-2022]	Same selection criteria as target trial & Age > 18 yrs [France, UK, 2007-2022]
Treatment	Iptacopan versus Anti-C5 therapy	Iptacopan	Anti-C5 therapy (eculizumab or ravulizumab)
Outcome(s) Intercurrent event	Hemoglobin Transfusion	Hemoglobin Transfusion	Hemoglobin Transfusion
Follow-up for outcomes (start/end)	Start of treatment until the end of study (24 weeks)	Time-0: initiate iptacopan Visit schedule: weeks (1,2, 4, 6, 8, 12, 16, 18, 20, 22, 24)	Time-0: initiate anti-C5 therapy Visit schedule: as-needed, expected every few weeks at anti-C5 injection

Modified from Supplementary Table S1 in Holt M et al 2025

Pitfalls: systematic bias, hypothetical examples

